

## Using agent-based models to generate transformation knowledge for the German Energiewende: potentials and challenges derived from four case studies

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# Using agent-based models to generate transformation knowledge for the German Energiewende – potentials and challenges derived from four case studies

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## Abstract

The German Energiewende is a deliberate transformation of an established industrial economy towards a nearly CO<sub>2</sub>-free energy system accompanied by a phase out of nuclear energy. Its governance requires knowledge on how to steer the transition from the existing status quo to the target situation (transformation knowledge). The energy system is, however, a complex socio-technical system whose dynamics are influenced by behavioural and institutional aspects, which are badly represented by the dominant techno-economic scenario studies. In this paper we therefore investigate and identify characteristics of model studies that make agent-based modelling supportive for the generation of transformation knowledge for the Energiewende. This is done by reflecting on the experiences gained from four different applications of agent-based models. In particular, we analyse whether the studies have improved our understanding of policies' impacts on the energy system, whether the knowledge derived is useful for practitioners, how valid understanding derived by the studies is, and whether insights can be used beyond the initial case-studies. We conclude that agent-based modelling has high potential to generate transformation knowledge, but that the design of projects in which the models are developed and used is of major importance to reap this potential. Well-informed and goal-oriented stakeholder involvement and a strong collaboration between data collection and model development are crucial.

**Keywords:** Energiewende; energy; transition; transformation knowledge; agent-based model

# 1 Introduction

Climate change is spurring the transformation of society and leading us towards reductions in emitted greenhouse gases (GHG), especially carbon dioxide (CO<sub>2</sub>). Governments around the world have become active to achieve lower levels of CO<sub>2</sub> emissions. This endeavor especially affects countries such as Germany where primary energy is heavily sourced from fossils. The German *Energiewende* is a deliberate transformation of an established industrial economy aiming at a nearly CO<sub>2</sub>-free energy system<sup>1</sup> while phasing out nuclear energy by 2023 (BMWi, 2017).

Over the last twenty years, the climate system, and therefore the necessary set of GHG reduction targets, have become much better understood. At the same time various energy models were developed in order to base long-term energy development scenarios illustrating potential pathways for the German energy system towards the ambitiously set CO<sub>2</sub> reduction targets (Hake et al. 2015). However, the established scenarios illustrate merely techno-economic pathways, designed around the optimal deployment of technical measures such as clean energy production technologies or energy efficiency measures.

The energy system is, however, a complex socio-technical system, characterised by fast and considerable changes of its structure, a high number of variables, an increasing decentralization of decision making among an increasing number of actors, and a rising complexity and numerous uncertainties (Bale et al. 2015; Pfenninger et al. 2014). Therefore, in order to transform it to a climate friendly or sustainable status, we need to not only consider technology and prices, but to focus explicitly also on the role of the actors involved (Pfenninger et al. 2014; Schot et al. 2016; Schubert et al. 2015; Sovacool 2014). Behavioural and institutional aspects must be given due consideration, as it is the actors themselves that make the decisions to adopt or to use the technologies that the scenario studies of the energy system focus on (Steg et al. 2015). The knowledge to do this is still underdeveloped due to focusing on techno-economic energy systems analysis in the past (Sovacool 2014).

Therefore “transformation knowledge” is limited to date, which is defined as “knowledge on how to shape and implement the transition from the existing to the target situation” (ProClim-1997). Transformation knowledge complements knowledge about clear targets and knowledge about the current state of the system (called “target knowledge” and “systems knowledge” respectively; ProClim- 1997). Transformation knowledge is concerned with established technologies, regulations, practices and power relations, and includes knowledge about different technical, social, political, economic, cultural, and legal methods and means that facilitate the transformation process from this current situation to the target situation (Pohl & Hirsch Hadorn 2007). Transformation knowledge is required to give actors on various levels orientation for taking action.

Holtz et al. (2015) have argued that models are useful tools to study processes of long-term fundamental change in socio-technical systems such as the energy system, because models are explicit, clear and systematic, allow for inferences about the dynamics of the system, and facilitate systematic experiments that might not be possible otherwise. Köhler et al. (2018)

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<sup>1</sup> Defined as 80 – 95% reduction of CO<sub>2</sub> emissions by 2050

have reviewed a range of existing modelling approaches with regards to their ability to address a number of key characteristics of such transitions.

We expect that agent-based modelling (ABM) is a method that can contribute well to exploring transformation knowledge since it provides the opportunity to include boundedly rational and socially influenced behaviour of actors into simulation models in a bottom-up approach. It therefore enables explorations of the future course of complex systems like the energy system, while considering that this future course is influenced by many behavioural and institutional aspects. This complements other bottom-up scenario-driven approaches for describing the energy future, such as normative and descriptive scenarios.

While existing studies (see next section) established the potential to study aspects of the Energiewende with agent-based models and have produced valuable insights, the overall impact of ABM on the Energiewende remains negligible up to now.

In this context, our aim in this paper is to carve out requirements for and characteristics of agent-based model studies to produce transformation knowledge in the context of the Energiewende. We propose that such insights might contribute to larger impact of ABM in the future.

We will investigate these issues through an in-depth analysis of the experiences from four recent applications of agent-based models on transformation processes related to the German Energiewende. Those models were developed within three research projects that were part of a research programme that aimed at gaining transformation knowledge in the sense described above. We therefore presume that these model projects constitute a valuable set of cases to draw on for achieving our research aims.

In our paper, we briefly outline existing work of ABM applied to energy systems before developing an analytical framework for the analysis of the four models. The analytical framework is then applied to each of the models. Based on the results, we discuss the potential and challenges of ABM studies to generate transformation knowledge and carve out requirements for and characteristics of ABM studies that are crucial for producing transformation knowledge before giving some final conclusions.

## **2 Brief review of ABM of energy systems**

Sensfuß et al. (2007), Pfenninger et al. (2014) and Ringler et al. (2016) provide reviews that include or focus on agent-based models of the energy system or relevant subsystems. Sensfuß et al. (2007) review agent-based modelling of the electricity markets. They conclude that agent-based simulations had developed from simple models to ones being capable of dealing with multiple markets and time scales, and that despite remaining challenges agent-based simulation appears promising to be useful as a test bed for the electricity sector. Pfenninger et al. (2014) conclude that methods from complexity science (including ABM) have been successfully applied in energy research, but that a need for a deeper and more fundamental treatment of complexity remains with much potential for additional research. Ringler et al. (2016) review the use of ABM for analysing smart grids from a systems perspective, which is still a limited field of research. They assess ABM to be a valuable approach to analyse opportunities and challenges in smart grids and markets. But they also warn that, when addressing new issues with explorative research designs, model calibration and validation are

particularly demanding, and that harmonization and consistency are essential to increase comparability between models.

Much analysis of energy markets and systems using agent-based models initially focused on wholesale markets, and did not consider social aspects of the energy system explicitly (Bagnall & Smith 2005; Bower et al. 2001; Bunn & Oliveira 2003). Agent-based models that include social science insights to study energy markets, energy technologies and energy systems in general are relatively recent developments, but are increasingly recognized for being supportive for policy design and evaluation as well as system design and infrastructure planning (Rai & Henry 2016). Although being a recent development, they can draw on experiences and insights from the broader social simulation field that has addressed aspects such as learning, opinion dynamics, social norms and the diffusion of innovations (Hauke et al. 2017; Kiesling et al. 2012; Meyer et al. 2009).

Until now only a few of studies using ABM with an explicit focus on the German energy system have been carried out. Early studies focus on the newly liberalised market: (Bower et al. 2001) used an agent-based model to analyse the effect of power generators' strategic behaviour on prices in the liberalised electricity market. Scheidt (2002) aimed to carve out the factors that influence volatility of spot-market prices. Müller et al. (2007) focused on increasing consumer prices in the German retail market. With an increasing share of renewable electricity and concerns about climate change, studies turned to investigate those issues: using the PowerACE model, Sensfuß et al. (2008) have studied the price effect created by renewables on the German electricity market, and Weidlich et al. (2008) studied the effect of CO<sub>2</sub> emissions trading on the electricity market. Reeg et al. (2012) analysed support schemes for the integration of renewable electricity into the electricity system, focussing on power-plant operators and intermediaries. More recently agent-based models have also been used to investigate spatial aspects of the energy system: Sorda et al. (2013) studied the effects of support schemes on the diffusion of and electricity generation from combined heat and power biogas plants. Lauf et al. (2015) assessed the spatial allocation of renewable energy sources under different policy scenarios. Ernst & Briegel (2016) and Krebs (2017) used a spatially explicit and psychologically grounded agent-based model to study the diffusion of green electricity in Germany.

### 3 Analytical framework

Up to date there does not exist an established framework for identifying whether models help creating transformation knowledge or not. Therefore, in order to carve out what has to be taken into account when designing model studies that shall create knowledge on how to shape and implement the transition from the existing to the target situation, we first have to further clarify the concept. To specify the term and concept "transformation knowledge" more precisely we developed a framework that considers four issues that are of particular interest:

1. Transformation knowledge includes being able to link interventions to (expected) impacts. Assessing the impact of interventions in complex systems such as the energy system is, however, challenging. We therefore investigate the *explanatory approach* of the selected agent-based models; i.e., how they (aim to) improve our understanding of policies' impacts on the complex energy system.

2. In order to unfold transformative impact, the knowledge derived from agent-based models needs to be usable by practitioners who are responsible for implementing policies and other measures to foster the energy transition. We therefore assess the *usefulness* of knowledge derived from the selected agent-based models for practitioners.
3. Conforming to requirements of “evidence based policy” or more nuanced demands for “good governance of evidence” (Parkhurst 2016) policy makers, who are accountable to citizens, need to be informed by transformation knowledge that is scientifically valid. We therefore investigate the *validity* of understanding derived by the selected agent-based models.
4. Knowledge derived from agent-based models unfolds larger transformative impact if insights are of a more general nature or transferable from the initial case-study to other cases. We therefore evaluate the *generality* of insights derived from the selected agent-based models; i.e., whether these insights can be used beyond the initial case-study.

As a next step we operationalize these four issues for application to the case studies. We discuss in the following these four issues in turn in the context of ABM more generally, thereby also providing a brief introduction to ABM for readers who are not familiar with it. While doing so, we briefly outline what each issue means in the context of the Energiewende; and based on that we specify the information that needs to be provided for each model study to assess its ability to generate transformation knowledge.

### 3.1 Explanatory approach

ABM allows for the study of complex systems from a bottom-up perspective and has been applied in a broad range of disciplines, ranging from molecular self-assembly, biology and ecology to economics, sociology, anthropology, and cognitive science (Macal & North 2010). Agent-based models define elements – the agents – that interact with each other and their (passive) environment. Agents thereby may be heterogeneous in their attributes and behavioral rules. When applied to social systems, at their core, agent-based models consist of agents that represent actors that interact with each other and their environment such as technical artefacts. Simulation of agents’ interactions over time allows for generating emergent phenomena on the level of a group, organisation or other collection of actors – be they spatial or temporal patterns or characteristic statistical distributions of variables of interest – based on the boundedly rational behaviour and social interactions of these actors (Gilbert & Troitzsch 2005; Heckbert et al. 2010; Hedstrom 2005). As such, ABM provides a means to identify possible explanations for emergent system-level behaviour by identifying underlying mechanisms (Mayntz 2004), i.e., by providing a causal reconstruction of the processes that account for the emergent phenomenon (Epstein & Axtell 1996). Policy impacts can thus be interpreted in ABM studies as emergent effects from policies through their effects on the behaviour and interactions of actors.

Understanding the impact of interventions in the complex energy system requires not only a consideration of the type and strength of the intervention, but also of the state of the targeted system. To outline the explanatory approach of the agent-based models presented in the next section for understanding of the impacts of interventions for shaping and implementing the



energy transition, we will describe three aspects: (1) the emergent phenomenon of interest that each model aims to explain; (2) the system representation including its main elements and processes, and how these elements and processes are involved in mechanisms that generate the emergent phenomenon of interest; (3) interventions that can be analysed with the model and how these affect the actors.

### **3.2 Usefulness**

It is central for the usefulness of results for practitioners that the model answers a question that is relevant for them. For gaining transformation knowledge, relevant questions are typically concerned with the effects of interventions; in particular, which (type of) intervention is most adequate to support the transition from the existing to the target situation.

Models in general, and agent-based models in particular, are built for a purpose and can not (easily) be used for a different purpose. It is therefore essential to make sure that the model purpose matches with the problems of the target group of stakeholders. This is typically best achieved through involvement of stakeholders in the model design process (Bale et al. 2015; Voinov & Bousquet 2010). The flexibility of the ABM method thereby facilitates a vast range of model designs, i.e. the method does not per se impose limitations to tailoring the model purpose to stakeholders' needs. There are, however, delicate interdependencies between a model's purpose, its design and resulting complexity, data availability, and validity. This makes some conclude that "developing good multi-agent models is still something of an art, rather than a science." (Gilbert 2004, p.8.). Consequently, agent-based models can not answer each and every question of practitioners to any degree of precision. Instead, the models' purpose and design need to find a match between practitioners' needs and what is manageable from the modeler's point of view.

The communication of model results to relevant stakeholders that are able to implement the proposed measures is of central importance for knowledge gained during a project to become transformation knowledge which is useful for practitioners. Communicating simulation results to stakeholders, however, is difficult in a variety of ways. Models include a lot of knowledge from various sources and many different assumptions, all of which influence (to various degrees) the model results. To fully understand a model's results therefore requires devoting a considerable amount of time to the study of the model itself. Limited understanding of the model may reduce the trust of stakeholders in the model, especially if the results do not match their expectations.

At least two different strategies exist to mitigate these difficulties. Some researchers have developed agent-based models together with stakeholders, and model design and the whole modeling exercise have been tailored towards a close collaboration with stakeholders (e.g., Barreteau et al. 2003; D'Aquino et al. 2003; Feuillette et al. 2003). This requires extensive time commitments and engagement from stakeholders, and the resulting models have to remain modestly complex. In such cases, the stakeholders might fully understand the model and (ideally) are willing to accept its results as they understand how they were derived. If agent-based models become large and complex software modules, in order to represent large and complex systems to a sufficient degree of detail, it can not be expected that stakeholders fully understand the models, nor that their results are readily accepted. A strategy for nevertheless communicating results from a model is to refrain from explaining the full model,

and to instead extract core mechanisms that generate interesting results, forming isolated explanations of those mechanisms and their consequences within the broader context of the model.

For agent-based models to produce useful transformation knowledge in the context of the Energiewende they hence need to assure that their object of research is of interest to stakeholders capable of acting on the energy system. They need to represent interventions and their effects; and it must be possible to communicate the results of these studies to the stakeholders who are able to implement the respective interventions.

In order to evaluate the usefulness of the models presented in the next section, we therefore: (1) analyse the relevance of the studied emergent phenomenon for the Energiewende; (2) outline whether and which interventions have been made explicit in their design and (expected) effects; and (3) discuss how the results have been communicated to practitioners with the powers to implement those interventions.

### **3.3 Validity**

Agent-based modelling research started with abstract models to showcase and explore fundamental principles (e.g. Epstein & Axtell 1996). This established its potential (e.g. Farmer & Foley 2009). Since then it has increasingly been recognized that empirical embeddedness of agent-based models, i.e., calibration and validation of models based on empirical data, matters for drawing policy relevant conclusions. Traditionally, validation is understood as testing whether the model captures reality sufficiently well through comparison of the model and its results with empirical data (Ormerod & Rosewell 2009; Windrum et al. 2007).<sup>2</sup> This includes to demonstrate the correspondence between model entities and processes that constitute the micro-foundation of the model with real-world entities and processes. It furthermore includes to demonstrate the ability of the model to replicate emergent empirical patterns on the system level. Conventional validation is, however, often challenging for agent-based modelling. Degrees of freedom for designing agent-based models are large, and knowledge is often limited when it comes to specific questions of model design, or might even be contradictory if different schools of thought propose different theories. Furthermore, social science theories often require a “translation” or interpretation when it comes to the details of implementation in the formal model language, which adds additional ambiguity and subjectivity. A certain level of subjectivity with regards to the model structure chosen by the model builder therefore is unavoidable for ABM. A lack of available data to parameterize the model often results in a number of free parameters, which can lead to the effect that a broad range of behaviour can be produced with an appropriate choice of parameter values. Therefore, empirical data that matches output variables of the model is typically needed for calibrating free parameters, and often little or no independent data is left for rigorous validation.

Scholars have argued, that validation should be more broadly understood as the process of assessing the usefulness of a model for its intended purpose, and that this may be achieved in

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<sup>2</sup> Validation is thus differentiated here from verification: the process of ensuring that the model performs in the manner intended by its designers and implementers (e.g., the program code does not contain errors) (Galán et al. 2009).

different ways than the conventional validation perspective proposes (e.g., Moss 2008). Over time, a range of conventional to more innovative strategies for calibration and validation have been developed that are sensitive to the type of model developed and its intended use (Grimm 2005; Hales 2003; Moss 2008; Windrum et al. 2007). Those approaches require either data from a set of comparable cases, historical time series, the involvement of stakeholders and experts, or compare models to each other.

We argue that modeling the Energiewende is indeed at the most challenging edge of the spectrum of calibration and validation, as we are not concerned with historical developments but the (unknowable) future, and the structure of the energy system changes at a fast pace which entails that historical data from a structurally similar system is not always available. Furthermore, the Energiewende is most advanced in Germany compared to other countries worldwide, and therefore no comparable cases exist that could be referred to (Quitzwet al. 2016). Furthermore, as outlined in the introduction, only a small number of models exist so far that could be used for comparison. As a consequence, involvement of experts and stakeholders as well as own data collection become important strategies for model calibration and validation.

To assess the validity of the insights generated by the models presented in the next section, we will discuss: (1) the overall approach adopted for calibration and validation, i.e., how it is assessed whether the model is fit for its purpose, (2) the theoretical basis, (3) the empirical data, and (4) the involvement of stakeholders.

### **3.4 Generality**

For social science in general and for ABM in particular a tension exists between understanding a particular case well and covering a wide range of empirical phenomena that share some common features. In order to describe this tension, (Boero & Squazzoni 2005) have distinguished between “case-based models” and “typifications”, while (Janssen & Ostrom 2006) introduce the dimension of “context” versus “generalization”. The ABM method does not impose any restrictions concerning the level of abstraction of the analysis. Which abstraction level is particularly useful for generating transformation knowledge is still an open question – i.e. it is debated which amount of “context” needs to be included in models for producing practically relevant knowledge and what can be represented in a more abstract way for the sake of better transferability (Edmonds 2010). Furthermore, agent-based models can be developed for a variety of empirical scales, ranging from small groups up to the supra-national level, and is therefore applicable to a broad range of research questions that relate to various empirical scales. As outlined above, we use the hypothesis, that the possibility for generalization of results from ABM studies beyond the initial case study helps to unfold larger transformative impact for the Energiewende.

To assess the possibility to transfer insights from the models presented in the next section to other, similar cases or the Energiewende as a whole, we check (1) the models’ empirical scale and (2) their level of abstraction. We (3) discuss the transferability of insights through carving out in which way the details of context considered in the model (if any) matter for the results.

## 4 Case studies

In this section we apply the analytical framework developed in the previous section to four agent-based models that were developed within the three projects “EnerTransRuhr”, “SW-Agent” and “Resysstra”<sup>3</sup> that were part of the research programme “Environmentally and socially compatible transformation of the energy system”<sup>4</sup> funded by the German Ministry for Education and Research (BMBF) within its long-term funding priority Social-Ecological Research (SÖF) as part of its Research for Sustainable Development (FONA) framework program.<sup>5</sup> Each project ran for three-years. For each model, we give a brief introduction and discuss the four issues identified Tables 1-4 at the end of each subsection summarise and assess each case.

### 4.1 Case 1 (insulation): Increasing homeowners’ wall insulation activity

In the EnerTransRuhr project, an agent-based model was developed in order to (1) increase the understanding of homeowners’<sup>6</sup> insulation activity (systems knowledge) and to (2) assess current and novel approaches aiming at an increase of homeowners’ insulation activity (transformation knowledge). The model is described in detail by (Friege 2016). It builds on a previous version (Friege et al. 2016) and a systematic secondary literature review (Friege & Chappin 2014).

*Explanatory approach:* The emergent phenomenon of interest generated by the model is a spatially and socially disaggregated scenario of homeowners’ wall insulation activities for a period of ten years. The model’s main actors are the homeowners; their decision-making process being at the core of the model. Homeowners are spatially distributed in a virtual city, and connected through social networks that are defined based on social homophily and spatial distance. To represent the decision-making of the homeowners, a logistic regression model was derived from empirical data. The main explanatory variables of this logistic regression model are: period of occupancy, attitudes towards insulation, the structural condition of the walls, and the homeowner’s age. Homeowners move out of buildings stochastically dependent on their current age, and are replaced by newly initialized homeowners. When a new homeowner moves into a building, the period of occupancy is set to zero. Attitude formation is assumed to be prone to social influences from the network of social contacts. The structural condition of the wall declines over time, while the homeowners’ age increases. Overall, the spatially and socially disaggregated scenarios of interest emerge from the homeowners’ decision-making process, their socio-geographic distribution, their social network, and the structural conditions of buildings at the beginning of the simulation. The model can be used to study the emergent effects of interventions that interfere with susceptible variables of the decision-making process of homeowners. The potential effect of providing information was studied by improving the attitudes of modeled homeowners towards insulation. A regulatory instrument was simulated by obligating homeowners who newly move into a house to carry out wall insulation within one year.

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<sup>3</sup> EnerTransRuhr”: <http://wupperinst.org/p/wi/p/s/pd/464/> , “SW-Agent”; <http://www.sw-agent.de/> , “Resysstra”: <http://www.resysstra.de/>

<sup>4</sup> „Umwelt- und gesellschaftsverträgliche Transformation des Energiesystems“, <http://www.fona.de/de/15980>

<sup>5</sup> Social-Ecological Research (SÖF), <https://www.fona.de/en/society-social-ecological-research-soef-19711.html>

<sup>6</sup> “Homeowners” thereby refers to owner-occupier households in detached and terraced residential buildings.

*Usefulness:* Several studies, exploring possible ways for reaching temperate climate countries' climate protection targets through the development of climate protection scenarios recommend increased efforts towards improving existing buildings' insulation (European Commission 2010; Kemp 2010; Taylor 2010). The recommendations were accepted and partly picked up by some governments' energy policies through including corresponding targets in their national energy concepts and introducing policy measures aiming at achieving those targets. However, currently there is a large gap between the needed and the current wall insulation rate. Policymakers have three options for increasing homeowners' insulation activity: economic means, information and regulations. According to (Friege & Chappin 2014), the empirical data confirms that homeowners' financial resources play no decisive role in their decision to carry out insulation projects. This casts doubt on the usefulness of widely implemented financial incentives aiming at an increase of homeowners' insulation activity, and highlights the importance of complementary interventions based on information and regulation. The potential of information instruments was explored in the simulation model through improving the homeowners' attitudes towards insulation. The results show that the potential is comparatively low, whereas it is advantageous – given a certain overall maximum effort – to focus the policy over a short period of time and to distribute the effort over many homeowners. According to model results, tightening regulatory instruments seems to be the best option: obligating new homeowners to insulate the wall within the first year after moving in would step into the window of opportunity associated with the change of house ownership, because many homeowners take advantage of this occasion to conduct extensive renovations anyway, but today many of those do not combine this with energetic renovation. Obligating new house owners to insulate walls within the first year after moving in has the potential to increase the total insulation rate by 40% in Germany for at least ten years (simulated time horizon). This does not yet correspond to doubling today's insulation rate of about 1% per year needed to achieve the German government's target of an 80% reduction in energy demand in the housing sector by 2050, but it would constitute a major step towards achieving that goal.

The structure and results of the model were presented to and discussed in the course of a workshop with representatives dealing with activities designed to increase the energy efficiency of buildings in the cities Bottrop, Dortmund and Oberhausen. Since the research has been issued by national ministries, we further expect that its results will be fed into the decision-making process on designing future policies aiming at an increase of homeowners' insulation activity.

*Validity:* The overall approach for calibration and validation followed was deriving the micro-foundation of the model from existing literature and in particular from empirical data; and calibration of the model to fit Germany's current annual wall insulation rate. A systematic secondary literature review (Friege & Chappin 2014) revealed existing knowledge on the decision-making process of homeowners, and identified potential research gaps. The results show that literature on energy-efficient renovation “lacks a deep understanding of the uncertainties surrounding economic aspects and noneconomic factors driving renovation decisions of homeowners” (Friege & Chappin 2014, p. 196). An online survey with 275 usable replies was carried out to fill these knowledge gaps. Regression analysis of the survey results revealed the above mentioned main explanatory variables. The analysis indicates that

knowing these variables is sufficient for correctly predicting about 80% of homeowners' wall insulation decisions. Parameter values for social network construction were derived from the empirical data and according to (Holzhauer et al. 2013). Attitude formation through social influence was modelled based on the results of the online survey and Latane's theory of social impact (Latane 1981).

For model parameterization, geospatial information system (GIS) data on the location of houses in the city of Bottrop, a city in west Germany, was used. The data provided also included information on each building's construction year. As older buildings are less likely to be already insulated, the buildings' construction year was used to estimate their insulation status at the beginning of the simulation. Based on findings from the online survey, a direct relation between the homeowners' attitude towards insulation and the insulation status of their buildings was assumed and used to distribute the attitudes towards insulation among the modeled homeowners. The homeowners' age was distributed according to overall statistics of Germany. The homeowners' overall level of renovation activity in the simulation model was calibrated to fit Germany's current annual wall insulation rate of 1.4% per year (Diefenbach et al. 2010). This was necessary 1) due to the generally higher renovation activity of the homeowners taking part in the survey, whose activity was used to derive the decision-making models, and 2) given the fact that Bottrop's existing building stock is older than the German average.

*Generality:* The empirical scale of the model is that of a city. The model is based on a worldwide literature review and was refined with data collected within an empirical study conducted in Germany. Information on buildings' ages and locations is specific for the city of Bottrop, a city in Western Germany located in the Ruhr Area. In particular, Bottrop's existing building stock is older than the German average. Information on homeowners was derived from national data. The model was calibrated to fit Germany's current annual wall insulation rate. The model parameters that stem from the city of Bottrop (buildings' ages and locations) have some effect on the magnitude of results, but not on the mechanisms identified and the qualitative effects of the analysed interventions. We therefore assume that the main conclusions regarding homeowners' decision-making processes and recommended approaches to increasing their insulation activity are transferable to other cities in Germany and temperate climate countries such as in northern Europe.

<i>Explanatory approach</i>	Spatially and socially disaggregated scenarios of homeowners' insulation activity emerge from homeowners' decision-making, their socio-geographic distribution, their social network, and the structural conditions of buildings.
<i>Usefulness</i>	The model has clear policy messages for the problem of increasing homeowners' renovation activity in Germany, which is of high relevance for the increase of energy efficiency, a core strategic objective of the Energiewende (BMWi 2017). The communication of the results to stakeholders has however been marginal.
<i>Validity</i>	The model has a sound micro-foundation; in particular, the core of the model, the decision-making process of homeowners, was derived from a broad review of empirical knowledge and from empirical data. The model could be calibrated to replicate historical data. However, no validation with independent system level data was possible, and stakeholders have not been involved in model design and

	validation.
<i>Generality</i>	The main conclusions regarding homeowners' decision-making processes and recommended approaches to increasing their insulation activity are transferable to other cities in Germany and temperate climate countries

**Table 1: Overall summary and assessment of case 1 (insulation) with regards to explanatory approach, usefulness, validity and generality**

## **4.2 Case 2 (retail market): The effect of regional preferences on the electricity retail market**

The second case study was conducted in the SW-Agent project and analyzed the retail market for electricity in Germany. The project highlighted the effect played by regional preferences of households which lead them to purchase from nearby suppliers. ABM was chosen as a methodology in the project potentially suited to studying the apparent preferences of households for local suppliers in addition to their preference for low prices. The model is described in detail by (Yadack et al. 2016a).

*Explanatory approach:* The emergent phenomena of interest are the dynamics of the composition of the electricity retailer population, of sales margins, and of the electricity prices for households over time. The model simulates competition for household customers by electricity companies on a computer generated landscape. The generated landscapes include rural, urban and metropolitan regions and associated population densities. Both utility companies and households are decision-making agents in the model. Households are imbued with separate preferences for price and regionality (represented by geographic distance) as well as a threshold for becoming dissatisfied with their electricity provider, and choose their preferred electricity company among those who offer in their region. Multiple electricity retailers compete in the simulation to win contracts to deliver electricity to customers by adjusting their tariff price, and through focusing expansion of their activities on relatively dissatisfied clusters of households. For the latter, electricity retailers choose to locate new generation and sales activities close to those relatively unsatisfied consumers, offering a regional electricity product. Explanatory variables in the model are, in addition to these agent-specific values for price and preference-weightings, the number of firms and households, and their geographic distribution (e.g., rural versus urban) in a given simulation run. Different simulation runs produce different development scenarios, depending on these influences and on initial values of market prices. The model can be used to study the interaction between the effects of geographic distribution of firms and households in communities and the regional preferences of households. The regulatory instrument first implemented and studied in the model was the liberalization of the retail market and the response of the household and firm agents to this regulatory instrument.

*Usefulness:* In a freely competitive retail market, electricity users are free to choose any electricity supplier, independent of the firms' and power plants' physical locations. In such a market, competition is expected to be intense, as electricity suppliers all sell a perfectly replicable and equivalent commodity. Through economies of scale and the need to leverage digital technologies, smaller, regional and in many cases community owned electricity utilities could be expected to be in danger of being priced out of the market entirely. Local utilities in Germany have, however, proven to be competitive despite competition from large

players, multinationals and internet-based new entrants to the sector. According to the German Association of Energy and Water Industries, since liberalization of the German retail market in 1998 up to 2015, prices for electricity (excluding taxes and surcharges) have increased for households by 7%. The same value for industrial electricity customers has decreased by 23% (Yadack et al. 2016a). This indicates asymmetric benefits of liberalization for the household electricity users and industrial electricity users, and in particular, leaves open the question as to which factors mediate the intensity of retail competition in the household electricity market when competition has so clearly intensified in the retail market for industrial electricity customers. The results of Monte Carlo studies with the developed agent-based model showed that regional preferences play a significant role in mediating the intensity of competition between electricity suppliers in a liberalized retail market. Sales margins calculated into electricity tariffs increased with increasing regional preferences. The development of these sales margins over time was also shown to be a non-linear process, and highly dependent on regional clustering of households and the threshold for dissatisfaction with one's electricity provider. Scenarios in the simulations highlighted not only the dynamics behind price developments, but also individual firms' strategies when their local customers fall into the scopes of low-price competitors. That local, municipally-owned electricity utilities could in the real world affect their own local residents' attitudes toward electricity (for example through diverting portions of profits toward community projects such as playgrounds and/or other community initiatives) suggests and supports the claim that these local utilities have been able to apply a certain degree of competitive advantage in regional and social factors to defend their positions as competitors in the retail electricity market – despite liberalization and intensifying competition in price. One useful lesson for policy makers is that social factors such as regionality can affect the development of market prices as shown by the model. Future policies aimed at increasing competition in electricity markets need to account for such factors if the response of markets is to be anticipated correctly. Failure to do so could result in even more difficult corrective measures later (e.g. discussion of subsidies and fee exceptions for large industry's electricity costs in Germany). In concluding experiences from stakeholder workshops, it could be argued that communication of the insights gained from the model to industry stakeholders was more effective than the communication to policy stakeholders. In particular, electricity retailers showed interest in practical applications of the model for individual case studies – this being due to their exposure to competitive retail market pressures which the model also simulates.

*Validity:* The main strategies for achieving validity were empirical data collection and the involvement of stakeholders to assure the correspondence of the model's micro-foundation with real world entities. The development of the model was imbedded in the larger scope of the project "SW-Agent", and as part of this project, was informed by both stakeholder workshops with utility companies and with empirical data from household surveys. To set up the model, a workshop of industry professions was organized to draft an initial structure of the model, which was then completed and implemented in close cooperation with the utility company. The structure of households' preferences in the model was inspired by a synthesis of work on national surveys of electricity customers of a utility company based in southern Germany (Yadack et al. 2016b). Over 4,000 households (from approximately 27,000 contacted households) responded to an online survey in January 2014. Respondents were



presented with fictive electricity tariffs in a choice-based analysis of their preferences and were asked to choose between various alternatives. The attributes of the presented tariffs included type of firm (including whether the firm was a local utility), renewable or conventional electricity, a level of service support, and price, and the respondents' answers were able to be collated with empirical data on location, and social demographic responses in the survey. The survey data showed a strong preference of households for local municipality-owned utilities. This granular empirical data was applied to the development of the abstracted simulation environment, and was used to structure and quantify preferences in the simulation. Population densities of different types of regions were based on public data from the German Federal Statistics Office. The pattern of suppliers' markups to electricity prices demonstrated in the model is reflected in empirical data on real prices in Germany since market liberalization in the electricity sector.

*Generality:* The model itself, and especially the way in which household agents make decisions, was based on empirical observations at the micro-level of German households, complemented by statistical data for Germany. The computer generated landscapes are generic. Although specific to Germany in its application and calibration, the structure of the model can apply to describing the effects of introducing a freely competitive, liberalized retail market for electricity into a region with a previously traditional regional-monopolistic market structure.

<i>Explanatory approach</i>	Dynamics of the composition of an electricity retailer population, of sales margins, and of the electricity prices for households emerge from the competition for household customers by electricity companies on a computer generated landscape. Households are imbued with separate preferences for price and regionality.
<i>Usefulness</i>	The model provides insight into the role of regional firms in the liberalized market and the effects and unintended side-effects of liberalization. It provides relevant insights for the goal of the Energiewende to safeguard competitiveness (BMWi 2017). The model also provides a sound basis to study combined effects of regionality and "green" electricity provided by regional electricity producers in future work. The model has raised the interest of electricity retailers for future practical applications.
<i>Validity</i>	The model has a strong empirical basis, in particular with regards to consumer preferences. The inclusion of stakeholders in model design enhances the validity of the model structure. The model was able to provide insights into empirical developments of retail prices and the markups underlying them. Validation with independent data was, however, not possible.
<i>Generality</i>	Although specific to Germany in its application and calibration, the structure of the model can apply to describing the effects of introducing a freely competitive, liberalized retail market for electricity into a region with a previously traditional regional-monopolistic market structure.

**Table 2: Overall summary and assessment of case 2 (retail market) with regards to explanatory approach, usefulness, validity and generality**

### **4.3 Case 3 (feedback products): Evaluation of marketing strategies for energy consumption changing feedback devices**

The third case study, which was also conducted in the EnerTransRuhr project, used ABM to

evaluate marketing strategies for feedback devices. Numerous approaches exist to give feedback to energy consumers, e. g. email, online platforms, or installed in-home displays (Darby 2006; Karlin et al. 2014; Laschke et al. 2011). One example of a feedback device is a so-called “CO<sub>2</sub> meter”, which shows the indoor air quality—measured by CO<sub>2</sub> level—via the colors of a traffic light. This was shown to be effective in convincing households to practice energy-efficient ventilation of rooms. In order to assist the large-scale leveraging of the potential of feedback devices, this case study addressed the question of which method of managing the rollout of feedback technology is most effective at creating additional energy-efficient heating behaviour, using the CO<sub>2</sub> meter as a proxy. The model, building on an earlier version (Jensen et al. 2016), is presented in detail by (Jensen & Chappin 2017).

*Explanatory approach:* The emergent phenomenon of interest is the spread of the CO<sub>2</sub> meter and of energy-efficient ventilation behaviour. The effect of the CO<sub>2</sub> meter is assumed to be the result of three interconnected processes: (1) the diffusion of the CO<sub>2</sub> meter through a population of households, (2) the direct impact of the CO<sub>2</sub> meter on ventilation behaviour of adopting households, and (3) a spreading of energy-efficient behaviour via social influence. The basic model components are household agents that are assigned to three different consumer groups that differ with respect to their attitude towards eco-friendly household products. Households are spatially distributed in a virtual city, and connected through social networks that are defined based on measured homophily and spatial distance. The agents make choices on whether to adopt the CO<sub>2</sub> meter and whether to adopt energy efficient ventilation behaviour. These choices are both modeled such that they can be influenced by other agents in their local social network. The model was used to assess strategies for a roll out of the CO<sub>2</sub> meter that differ with respect to initial attributes of the initial adopters (consumer group, centrality in social network).

*Usefulness:* Providing feedback to energy consumers about their energy consumption behavior can help them to save energy. Feedback about behavior was found to decrease energy consumption by up to 20%, with an average of about 10% (Karlin et al. 2014; Wood & Newborough 2003). Further, achieving this behavioural contribution to energy efficiency is comparably cheap and simple. In simulations, marketing strategies for feedback devices successfully resulted in additional adoption of energy-efficient ventilation behavior, particularly when: (1) the use of feedback devices was incentivized economically by giving away or lending out some devices for free; (2) the social influence of well-connected opinion leaders (i.e. households that are particularly influential for others) was leveraged to promote the devices; and (3) households of higher social status were targeted primarily by marketing (Jensen & Chappin 2017). The tested marketing strategies rank differently when compared based on their effectiveness than if compared for cost-efficiency. Regarding effectiveness, (Jensen & Chappin 2017) found the lending out of devices to be the best performing marketing strategy. The most cost-effective measure is raising awareness among households of higher social status. The issue of communicating simulation results was approached through identification of stakeholders on city level that are best able to reach the most promising target groups identified in the various strategies and who at the same time might be interested to implement the respective measure (e.g., lending out devices). Discussions with relevant stakeholders have been initiated towards the end of the project (when model results were available), but the process fizzled out in an early stage when the project ended.

*Validity:* The overall approach chosen for calibration and validation was pattern-oriented modelling (Grimm 2005), at which model parameters are accepted if model output matches simultaneously multiple defining characteristics of the observed behaviour of the real system. The Theory of Planned Behavior (Ajzen 1991) was used as theoretical framework to structure the actors' decision making and to combine intrinsic motivation and social influence. The behaviour diffusion process was parameterized from survey data (Jensen et al. 2016). As no data for the diffusion of feedback devices was available, data on the diffusion of water-saving showerheads (Schwarz & Ernst 2009) was used as a proxy; arguing that these are similar with respect to characteristic that are decisive for production diffusion (Jensen et al. 2016). The impact of feedback devices on the energy demand of adopting households was derived from data from field-tests of the CO<sub>2</sub> meter. The agent population (consumer groups) and the social network was parameterized using empirical data on households and buildings. Most of this data was gathered in the city of Bottrop (Germany). Using commercial high-resolution marketing data, about 32,000 modeled households were mapped to the 3 consumer groups that differ in their preferences of technology adoption.

*Generality:* The empirical scale of the model is that of a city. The model was calibrated to the city of Bottrop with regards to agents' consumer groups and local social network data. But, results of the models were shown to be generalizable to other German cities: a sensitivity analysis showed that model implications are specific to the composition of consumer groups in a simulated area. However, results are not sensitive to the detailed urban structure or the exact location where marketing is taken out (Jensen & Chappin 2017). The insights from the model are in principle transferable and applicable to other behavior-changing feedback devices; assuming their characteristics for diffusion being similar.

<i>Explanatory approach</i>	The spread of the feedback device CO <sub>2</sub> meter and of energy-efficient ventilation behaviour emerge from three interconnected processes: (1) the diffusion of the CO <sub>2</sub> meter through a population of households, (2) the direct impact of the CO <sub>2</sub> meter on ventilation behaviour of adopting households, and (3) a spreading of energy-efficient behaviour via social influence.
<i>Usefulness</i>	The model provides an assessment of strategies for a roll-out of feedback products through analysing the system level effect on energy savings. We conclude that the model results are useful for practitioners because they give clear advice for cheap and implementable options for increasing energy efficiency, which is a core strategic objective of the Energiewende (BMW 2017). The communication of the results to stakeholders has however been marginal, as there was no project time left when results were available.
<i>Validity</i>	The model has a solid micro-foundation and could be calibrated to historical data. However, validation with independent system level data could be extended, and stakeholders have not been involved in model design.
<i>Generality</i>	The model is calibrated to the city of Bottrop. Results are, however, shown to be not sensitive to the detailed urban structure or the exact location where marketing is taken out. The insights from the model are therefore transferable to other cities, and in principle applicable to other behaviour-changing feedback devices; assuming their characteristics for diffusion being similar.

**Table 3: Overall summary and assessment of case 3 (feedback products) with regards to explanatory approach, usefulness, validity and generality**

#### **4.4 Case 4 (power-to-fuel): Evaluation of pathways for an economic integration of power-to-fuel concepts**

In the fourth case study, which was conducted in the Resysstra project, an agent-based model was developed to evaluate regulatory pathways for an economic integration of renewable synthetic fuels into the German energy system (Schnülle et al. 2016). A full documentation of the model is provided by (Brand et al. 2017).

*Explanatory approach:* The emergent phenomenon of interest is the economic uptake potential of power-to-fuel applications. Power-to-fuel operators, customers and subsidizers were implemented as agents in a newly developed agent-based model. In order to represent the whole supply and demand volumes while keeping the model complexity manageable, each group of similar real-world actors was represented by a single agent in the model. These agents were set into relation with one another via their participation in assigned energy markets. In general, the model revolves around investment decisions made by the variety of operators, and purchasing decisions made by the individualized customer groups, all within a dynamic context of evolving regulation and technological progress. In addition, the model includes technology classes for the production of hydrogen, methane and liquid fuels, (installed) technology instances and business cases as passive objects. Finally, the regulatory and economic context for the interaction of agents and objects is represented in the model's environment. The model is able to simulate the effects of changes to economic variables such as taxes, fees and subsidies, as well as to rules and regulations.

*Usefulness:* The Energiewende process is increasingly facing challenges posed by the volatile fluctuation of electricity generation from wind and photovoltaic power. In accord, solutions for energy intensive applications like industrial high temperature processes and mobility concepts as aviation, shipping and heavy transport are required to achieve international climate protection targets. Power-to-fuel concepts represent an innovation path that might become the missing link as (excess) electricity can be utilized for a dynamic hydrogen and synthetic hydrocarbon production (Breyer et al. 2015; Graves et al. 2011). The technological readiness level of several power-to-fuel concepts allows for an upscaling progress to a large scale industrial application. However, due to high investment costs, disadvantageous framing conditions and low prices for fossil fuels the existing pilot projects are, except of small niche markets, not yet economically feasible and financially dependent on funding (Haarlemmer et al. 2014; Tremel et al. 2015). Based on stakeholder proposals four different options for political decision makers were considered to increase the market share of synthetic fuels and their effects were simulated with the model: (1) Increasing the cost for fossil fuels through the implementation of CO<sub>2</sub> taxes/high EU ETS<sup>7</sup> certificate prices; (2) less regulatory barriers for operators; (3) stronger subsidization of new projects in the early adoption phase; and (4) implementation of strong emission reduction quotas for all industrial sectors. The simulations revealed that even increasing market prices by up to 4% per year for fossil gas and oil hardly create business cases for power-to-fuel operators without any further regulatory interventions in the energy markets. Modifications of regulatory framework conditions in the national

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<sup>7</sup> EU ETS = European Union Emissions Trading System

energy laws, e.g. an exemption for operators from specific fees and taxes regarding input electricity for their production processes, appeared to be one of the key parameters for larger investments in power-to-fuel projects. However, the creation of future business cases that are independent from subsidization it requires strict emission reduction restrictions for industrial companies as this will raise their willingness-to-pay for CO<sub>2</sub>-neutral fuels in order to avoid penalties. A slow overall capacity development within the next decades could be the consequence of a lack of strong political climate protection measures.

The regulatory modifications analysed in this study have been part of debates between stakeholders and policy for years (Schenuit et al. 2016). Recent development indicates that political decision makers are considering modifications in the national energy law. However, for a large-scale integration of system innovations such as power-to-fuel and a consequent emission reduction in the overall energy system it requires additional strong political signals.

*Validity:* The main strategies for achieving validity were drawing on existing case-specific conceptual work and the involvement of stakeholders to assure the correspondence of the model's micro-foundation with real world entities. The successful integration of power-to-fuel approaches into energy systems is dependent on a variety of social, technical, economic and ecological influences. Hemmelskamp defines and structures such influencing interactions with the multi-impulse-hypothesis as "impact factors" (Hemmelskamp 1999). This hypothesis was used to identify and characterize important stakeholders that may influence the innovation process of power-to-fuel. Rational choice theory was chosen as a basic theoretical framework. For an explicit characterization of agents, extensive literature reviews were supplemented by expert knowledge gained in a series of interviews and workshops. In a total of three workshops with renowned researchers and industrial company representatives, information of stakeholders from the field was firstly structured in a mind map and the stakeholders were introduced into the method of ABM. Based on the revealed insights and open questions that appeared to be most prominent a basic model framework was constructed within a scientific workshop. In the third workshop, the model framework was discussed and further adjusted based on new insights from stakeholders once again. Finally, additional information e.g. on investment strategies was subject to discussions of researchers with individual stakeholders. Those sources highlighted that differentiated customer and operator groups as well as public subsidies were key to answer the research question. The simulations enabled for a quantitative assessment of complex interactions between decision makers and stakeholders who adapt to changing framing conditions. The model's results were further validated by the model's ability to replicate the power-to-fuel capacities in the real-case between 2009 and 2016.

*Generality:* The scope of the model is on a national level with regard to the German energy system. This is a logical consequence of the central model content, which covers a possible expansion of a system innovation such as power-to-fuel. To this end, it was necessary to implement energy and cash flows of energy markets, which extend beyond regional, smaller energy systems. Additionally, all legal decisions influencing the spread of power-to-fuel are implemented on a national level. However, geographic dispersion at the same time is not part of the model environment since local conditions have less impact at this early state of development, and this increases the degree of abstraction of the model. Some technological specification is also kept on an abstract level as e.g. there is no distinction between different

types of electrolyzers. These generalizations enable a better transferability of the generated knowledge as the information regarding markets and technologies might be easily replaced by new information for other systems.

<i>Explanatory approach</i>	The economic uptake potential of power-to-fuel applications emerges from the interactions of power-to-fuel operators, customers and subsidizers. In general, the model revolves around investment decisions made by the variety of operators, and purchasing decisions made by the individualized customer groups, all within a dynamic context of evolving regulation and technological progress.
<i>Usefulness</i>	Synthetic fuels are an important technological option to expand the use of renewable energies, what supports the respective strategic objective of the Energiewende (BMWi 2017). The model provides simulations of future synthetic fuel market shares in correlation to changing regulatory framing conditions. The impact of single policy adjustments on the capacity growth could successfully be demonstrated. Although it can be concluded that the results can be of high relevance especially for political decision makers, the results of the simulations were not discussed with stakeholders.
<i>Validity</i>	The basic setup and scenarios of interest were developed according to stakeholders' input and feedback, and the model was calibrated to a replication of historical data. Validation with independent system level data was not possible, and some parameter values remain unknown and add some degree of subjectivity to the results.
<i>Generality</i>	The model is tailored to the German energy system and takes a national level perspective. Some technological specification is kept on an abstract level what enables an easy replacement by new information for other systems.

**Table 4: Overall summary and assessment of case 4 (power-to-fuels) with regards to explanatory approach, usefulness, validity and generality**

## 5 Discussion of potential and challenges

Based on the experiences with the four agent-based models described above, we now discuss the potential of ABM to generate transformation knowledge, and the challenges for designing agent-based model studies to achieve this. We therefore follow the structure of our analytical framework.

### 5.1 Explanatory approach

ABM can improve our understanding of the energy system through providing a formal language to consistently integrate knowledge from various sources and disciplines. In particular, ABM allows for the inclusion of bounded-rational and socially influenced behaviour as well as institutional factors. Simulations then facilitate an enhanced understanding of emergent phenomena that are difficult to understand without considering these social factors; such as the low insulation rate in case 1 (insulation), and the competitiveness of small regional electricity utilities during market liberalization in case 2 (retail market).

Simulations furthermore allow to assess the relevance of each of a range of potentially important factors through studying their (emergent) effects. For example, in case 2 (feedback products), simulations allowed to investigate the interplay and the effects of product diffusion and behaviour diffusion. In case 4 (power-to-fuel), interdisciplinary knowledge (social,

techno-economic and regulatory impact factors) was firstly gathered in a qualitative model. However, with regard to the studied research question, the individual “magnitude of impact” of those single factors was completely unknown. Being translated all at once into an agent-based model enabled a quantitative evaluation of these impacts on the innovation system and in this way made it possible to provide guiding orientation for the transformation process.

Finally, the impacts of specific policies can be studied. Typically, simulations with agent-based models do not provide exact numbers for policies’ impacts, but they help to understand why certain policies unfold a stronger or different effect than others through making the underlying mechanisms operating in the system visible. For example, in case 1 (insulation) it became apparent that increasing homeowners’ information about and attitude towards wall insulation (alone) has limited effect as homeowners do not enter deliberation about renovation as long as they perceive the condition of their walls to be good enough. In case 4 (power-to-fuel), it could be shown that an increase of industrial companies’ willingness-to-pay for CO<sub>2</sub>-neutral fuels through emission reduction restrictions is required to achieve an economic uptake potential of power-to-fuel products that does not depend on subsidies.

## **5.2 Usefulness**

All the presented cases deal with topics that are relevant for the Energiewende and help to fulfill its goals and objectives. Case 1 (insulation), case 3 (feedback products) and case 4 (power-to-fuel) also study and compare the effects of various specific interventions that could be used to steer the Energiewende; while case 2 (retail market) in particular highlights the relevance of a sound systems knowledge to understand (and avoid) unintended side-effects of political measures (in this case: liberalization). In case 2 (retail market) and case 4 (power-to-fuel) stakeholder workshops were conducted to assure the match of the model purpose with questions of the stakeholders, and in case 1 (insulation) the result were presented to stakeholders on the local level.

Despite these efforts to assure the practical relevance of the research, there were difficulties to reach those stakeholders that are in charge to implement the proposed measures. The regulation suggested in case 1 (insulation) to introduce an obligation to insulate walls when moving to a new house, the design of liberalized power markets in case 2 (retail market) as well as the measures suggested in case 4 (power-to-fuel) (e. g., emission restrictions, fees and taxes) all need to be implemented by policy makers on national or even EU levels. Those stakeholders have not been involved in the projects, and therefore their awareness of the project results is likely limited. This is a main drawback of the analysed model studies regarding their transformative impact.

The main recipients of the results of these projects can be expected to be participants of the respective national discourses, who might pick up project results and make use of them during debates. Thus project results might influence policy making indirectly through providing arguments to be used in national discourses that precede policy making. In order to be useful for these discourses it becomes important that key messages of the models can be communicated and used independent of the model itself; i.e., that core results and mechanisms can be extracted from and comprehended without the model.

The following gives an example of such an abstraction from the model: In case 3 (feedback products) the model highlighted that effects of promoting feedback devices on the city level by the year 2030 are lower than can be expected by simply upscaling effects from single households through multiplication with the number of households in the city. The speed of dissemination of these products can robustly be estimated to be so slow that full penetration is not achieved by 2030. A large number of households already practice energy efficient ventilation behaviour, meaning that no additional effect can be expected in these households. Finally, it can be expected that energy efficient ventilation behaviour would also continue to spread (albeit more slowly) even if the CO<sub>2</sub> traffic light device was not introduced. These insights into the low effect at the city level became apparent through the integrated consideration of different factors in the model, but appear to be comprehensible also in the absence of a model. Similar storylines can be derived from the other models. However, this has not been done systematically to communicate model results to essential target groups.

### **5.3 Validity**

The flexibility of ABM facilitates the integration of knowledge from various theoretical and empirical sources, and prevents relevant knowledge from being neglected just because it is not within the frame of the chosen theoretical model, as might be the case for more rigid approaches. For example, in case 1 (insulation), the decision making process of homeowners was specifically designed to capture those characteristics of homeowners that appeared to be most relevant, based on an extensive survey of empirical studies. They are furthermore put in relation with the technical properties of different refurbishment options, a building typology, the current state of refurbishment, spatial structures of cities, and policy measures. In case 4 (power-to-fuel) the model was designed to depict specific interests of different industries which lead to representative investment decisions. A further benefit of the ABM approach for case 4 (power-to-fuel) was the possibility to include technical learning effects which correspond to global installation rates, and to evaluate their feedback on business cases. This was possible through representing all relevant market participants and the related consequences of their actions on other actors and techno-economic variables.

The cases presented above show that the principal uncertainty and subjectivity related to model design that arise from limited knowledge and ambiguity and subjectivity when choosing and operationalizing social science theories can not be erased. But it can be reduced through different practical strategies. Strategies to confirm that the model design considers the most relevant aspects include extensive literature surveys (case 1: insulation and case 4: power-to-fuel), discussions of model designs in interdisciplinary project teams (all cases), and stakeholder interviews and workshops (case 2: retail market and case 4: power-to-fuel).

The second main challenge for validity discussed in the analytical framework section arises from the fact that empirically based agent-based models are typically data intensive. In each of the four cases, both sources from the literature and own data collection (e.g., stakeholder dialogues, surveys, living labs) were used to gather the required information. Data availability and quality of the data has had a large influence on the models and their results, and data collection was directly aligned with the design of the models. Despite the intensity of data gathering efforts, some parameters could not be defined precisely for a range of practical reasons. Sensitive data (especially economic data) from companies is often not available, and



this lack of important data has to be compensated by assumptions which increase the error of the model results. In some cases, the gathered values for certain model parameters differed dramatically from data source to data source, which makes data more difficult to handle and interpret correctly. In case 4 (power-to-fuel), for example, there was a lack of sound investment figures for implemented technologies. An extensive literature survey showed that specific investment costs differed by up to a factor of 4 for the same technology, depending on the literature source or manufacturer information. In a major model variable, such inconsistencies may dramatically falsify the results. Thus, the modeller has to choose whether a source is more trustworthy than another one or to use average values. However, both choices are accompanied by significant uncertainties. The design of surveys meant to be used in modelling studies presents another particular challenge for ABM research as the model defines data needs but the information retrieved from the collected data may in turn influence the model design which then may lead to further data needs. For example, case 2 (retail market) focused on the importance of social preferences in the context of electricity retail markets and the model design was inspired by ex-post analysis of the survey results of utility customers and their preferences. These examples show, that the alignment of data collection and model development becomes of great importance for developing valid models.

The limitations arising from lacking data can (partly) be compensated through sensitivity analysis in which parameter values are varied and the effect of parameter variation on results is examined. In all case studies, some sensitivity analysis was conducted. For example, in case 2 (retail market), a sensitivity analysis was carried out to test the effect of both wholesale market prices and customer preferences on the level and development trends in retail market prices. In case 4 (power-to-fuel) a sensitivity analysis on the impact of future fossil energy prices (which determine demand potentials for alternative fuels but are highly uncertain) was applied to examine their direct influence on market integration potentials.

We conclude that ABM provides the possibility for valid model structures that include the most relevant variables and mechanisms of the represented socio-technical system. But in order to achieve this, a sound, systematic process of model building and testing including stakeholder involvement and empirical data collection is essential.

## **5.4 Generality**

As the cases illustrated, results from ABM studies are often generalizable and transferable to other geographical areas or to similar technologies. In all cases, some mechanisms were identified that are independent of the specific empirical application case (if any was chosen). For example, the mechanism that leads to a low insulation rate in case 1 (insulation) is not specific for the city of Bottrop. Additionally, in some cases, it became clear which parameters need to be adapted when repeating the evaluation at a different location or for a slightly different product. An example is case 3 (feedback products) in which the sensitivity analysis showed that quantitative results are sensitive to the composition of the consumer group, but not to the specific distribution of consumer types in the urban area. Data on the composition of the consumer group therefore needs to be collected and included in the model for applications of the model to different cities, while a detailed map of locations of consumer types is not required.

It also became clear during the case studies, that the extent to which insights are generalizable beyond the initial case study depends on the causal mechanism operating in the system, and the extent to which the most important explanatory variables and processes are idiosyncratic for the analysed case. This is not known beforehand, as understanding about causal mechanisms only develops during the project. But the ABM method facilitates clarification of the causal mechanisms and, through sensitivity analysis, allows for assessing which insights are transferable to cases beyond the initial case study.

## 6 Crucial aspects for production of transformation knowledge

The four discussed different agent-based models are not representative for all potential applications related to transformations of energy systems, and each model study has its unique merits and limitations. Yet, we consider this number high enough to draw some tentative conclusions on characteristics that are crucial for ABM studies to successfully produce transformation knowledge. Those are summarized in table 5.

Well-informed and goal-oriented inclusion of stakeholders
Possibility to refine model design based on new insights
Flexibility to adapt data collection to changing model designs
Several loops of stakeholder involvement and data collection
Adaptation of dissemination strategies to new system understanding
Long project duration

**Table 5: Recommendations for the design of projects with the goal to create transformation knowledge**

The above discussion revealed that agent-based models offer high potential as tools for generating transformation knowledge. Their ability to provide a bottom up representation of aspects of the complex socio-technical energy system based on empirical knowledge allows in turn for tailoring models towards inclusion of the empirically observed most important factors, for orienting models towards the questions that really matter for stakeholders, and for understanding policies’ impacts as result of changing actor behaviour. We have seen in the discussion above that to achieve this, the design of the model must be well aligned with stakeholder involvement and the collection of data. Therefore, we conclude that the design of projects in which agent-based models are developed and used is of major importance to reap their potential.

### 6.1 Stakeholder involvement

The identification of relevant target groups and the communication of project results to those groups is of major importance for the produced knowledge to unfold transformative impact. Ideally, the target groups for policy messages derived by the model exercise should be

involved in the project to ensure that the model answers a question that is relevant for these stakeholders and to increase understanding of the model and trust in model results. Stakeholder involvement is also important for the validity of models as practitioners with systems knowledge are typically indispensable for filling knowledge gaps in the scientific literature and for providing specific information about the target system. It is important to be aware that the specific stakeholders to be included for the different goals might differ; in particular stakeholders who can provide data might differ from those in charge to implement measures.

An important issue for project design therefore is the well-informed and goal-oriented involvement of stakeholders. The literature on participatory modelling (Voinov & Bousquet 2010), stakeholder involvement in sustainability science (Mielke et al. 2016), and more generally transdisciplinarity (Bergmann et al. 2012; Bernstein 2015) are helpful starting points to achieve this.

However, the cases show that planning the involvement of stakeholders during project design well is no guarantee for successful production of transformation knowledge, as the measures identified to be appropriate and therefore relevant target groups to be involved may change when knowledge of the case increases. For example, case 1 (insulation) initially set out to identify means to use networks between homeowners to foster insulation activity. This could have been done by local stakeholders involved in the project. But, increased understanding gained during the course of the project revealed that using networks, e.g. through information campaigns, has limited effects, and other measures (regulations) appear to be more effective. Those measures, however, can only be implemented by stakeholders on the national level (who were not involved in the project). Therefore, the possibility to adapt stakeholder involvement to increased understanding of the system would be ideal, from the modelling perspective.

## **6.2 Model calibration and validation**

Applicable data is essential for calibration and validation. A main observation of the case studies is that the understanding of the system develops during the course of the project, and therefore data needs could not be known exactly before or at the beginning of the project. This is in particular the case if stakeholders are involved in defining a model's purpose and therefore its structure. In this case, data collection cannot follow a predetermined design. Similarly, the relevance of stakeholders for knowledge acquisition might change during the course of the project. Flexibility in the project design to adapt data collection and stakeholder involvement to increasing understanding and changing model designs would therefore be ideal from the modeller's point of view.

A promising but resource intensive option for model validation and calibration is to couple simulation experiments with real life experiments. Real life experiments conduct interventions in a real system of a limited scope to study its effects (Claude et al. 2017; Liedtke et al. 2015; Schneidewind & Scheck 2013; Wagner & Grunwald 2015). One example would be a publicly funded programme for marketing and evaluating the use of energy consumption changing feedback devices as described in case study 3 (feedback products). In principle, virtual experiments with a simulation model can help to select most promising candidates for interventions in real life experiments, while the results from real life

experiments can validate the simulation model, or support its improvement. The merit of real life experiments over other methods of data collection such as workshops, interviews and surveys is that data for emergent properties of the complex socio-technical system can be collected. A major challenge is, however, that it takes a considerable amount of time (typically months or years) for the emergent effects to become visible, and long-term monitoring programs are time consuming and expensive beyond the usual project scope. Furthermore, the intervention to be studied is only one out of typically manifold influences on the system, and therefore isolating its specific effect is a challenge. Despite these challenges, we believe that the coupling of simulation experiments with real life experiments warrants further attention and trial.

### **6.3 Dissemination of results beyond the case study**

The assessment of transferability of case study results requires an understanding of the mechanisms operating in the model. This implies that transferability of results cannot be known exactly before or at the very beginning of the project, as the understanding of causal mechanisms operating in the system is limited at that point of time. It is therefore difficult for project designers to advocate for transferability of their “exemplary case” to other cases when applying for the project. Which target groups beyond the case study area should be targeted during the dissemination of project results is therefore difficult to plan when the project proposal is written. Ideally some flexibility to adapt dissemination activities to the understanding gained during the project should be allowed for in the project design.

### **6.4 Project duration**

Besides an increasing flexibility in project design, also the duration of the research project plays a large role in the context of ABM. Longer lasting projects, starting from 5 years, would have the benefit that stakeholders could be involved in several loops of model creation, which would help for purposeful design, calibration and validation of these models. Also real-life experiments require much time (see above), but would be very helpful, e.g. if the goal involves refining an already developed agent-based model with data from experiments.

## **7 Conclusions**

We have set out to identify requirements for and characteristics of agent-based model studies that support the generation of transformation knowledge for the Energiewende and for energy transitions. Agent-based models are in principle well suited to produce transformation knowledge. They allow for bottom up representation of the complex socio-technical energy system based on empirical knowledge. This allows for tailoring the model towards inclusion of the empirically observed most important factors, including boundedly rational behaviour of actors, social influence and institutions. The flexibility of the ABM method furthermore facilitates the orientation of model design towards the questions that really matter for stakeholders. The bottom up nature of agent-based models provides the possibility to identify potential policy impacts as emerging from their effects on the behaviour and interactions of actors. The transferability of insights to applications beyond the initial case study depends on the causal mechanisms operating in the system, and how specific these are for the modelled case. The ABM method helps to clarify this.

We conclude that the availability of systems knowledge and data is essential to achieve valid models, but the satisfaction of data needs of agent-based models is typically challenging. Well-informed and goal-oriented involvement of stakeholders and a strong collaboration between data collection and model development is most important to reap the potential of ABM in real-world projects. Therefore, the design of projects in which agent-based models are developed and used is of major importance for the success of modelling exercises. We argue that the “ideal project” involves well-informed and goal-oriented inclusion of stakeholders, and provides flexibility to adapt data collection, the inclusion of particular stakeholders and dissemination strategies for increasing understanding of causal mechanisms operating in the system and changing model designs. Projects with a duration of five years or longer that allow for involvement of (potentially changing) stakeholders in several loops and for conducting real life experiments in interaction with simulation experiments would be most beneficial.

As of yet, research using ABM has not become mainstream, despite its relevance in studying crucial elements typically underemphasized by the more traditional approaches for energy system modelling. Facing the huge challenges regarding actors and their behaviour, emerging not only from the Energiewende, but also for a sustainable transition in other sectors like mobility, industry or agriculture and the interconnections between them, would justify including calls for ABM in future research programmes.

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## References

- Ajzen, I. (1991). The theory of planned behavior. *Organizational Behavior and Human Decision Processes*, (50), 179–211.
- Bagnall, A. J., & Smith, G. D. (2005). A multiagent model of the UK market in electricity generation. *IEEE Transactions on Evolutionary Computation*, 9(5), 522–536.
- Bale, C. S., Varga, L., & Foxon, T. J. (2015). Energy and complexity: New ways forward. *Applied Energy*, 138, 150–159.
- Barreteau, O., Antona, M., D’Aquino, P., Aubert, S., Boissau, S., Bousquet, F., Mathevet, R. (2003). Our companion modelling approach. *Journal of Artificial Societies and Social Simulation*, 6(2), 1. Retrieved from <http://jasss.soc.surrey.ac.uk/6/2/1.html>
- Bergmann, M., Jahn, T., Knobloch, T., Krohn, W., Pohl, C., Schramm, E., Faust, R. C. (2012). *Methods for transdisciplinary research: a primer for practice*. Campus Verlag.
- Bernstein, J. H. (2015). Transdisciplinarity: A review of its origins, development, and current issues. *Journal of Research Practice*, 11(1), 1.

- BMWi (2017). A target architecture for the energy transition: from policy goals to specific measures. Retrieved 24 January 2018, from <https://www.bmwi.de/Redaktion/EN/Artikel/Energy/target-architecture.html>
- Boero, R., & Squazzoni, F. (2005). Does Empirical Embeddedness Matter? Methodological Issues on Agent-Based Models for Analytical Social Science. *Journal of Artificial Societies and Social Simulation*, 8(4), 6. Retrieved from <http://jasss.soc.surrey.ac.uk/8/4/6.html>
- Bower, J., Bunn, D. W., & Wattendrup, C. (2001). A model-based analysis of strategic consolidation in the German electricity industry. *Energy Policy*, 29(12), 987–1005.
- Brand, U., Giese, B., von Gleich, A., Heinbach, K., Petschow, U., Schnülle, C., Wigger, H. (2017). *Auf dem Weg zu Resilienten Energiesystemen! Resiliente Gestaltung des Energiesystems am Beispiel der Transformationsoptionen EE-Methan-System und regionale Selbstversorgung* (Final project report RESYSTRA). Universität Bremen.
- Breyer, C., Tsupari, E., Tikka, V., & Vainikka, P. (2015). Power-to-gas as an emerging profitable business through creating an integrated value chain. *Energy Procedia*, 73, 182–189.
- Bunn, D. W., & Oliveira, F. S. (2003). Evaluating individual market power in electricity markets via agent-based simulation. *Annals of Operations Research*, 121(1–4), 57–77.
- Claude, S., Ginestet, S., Bonhomme, M., Moulène, N., & Escadeillas, G. (2017). The Living Lab methodology for complex environments: Insights from the thermal refurbishment of a historical district in the city of Cahors, France. *Energy Research & Social Science*, 32, 121–130.
- D’Aquino, P., Le Page, C., Bousquet, F., & Bah, A. (2003). Using self-designed role-playing games and a multi-agent system to empower a local decision-making process for land use management: The SelfCormas experiment in Senegal. *Journal of Artificial Societies and Social Simulation*, 6(3), 5.
- Darby, S. (2006). *The effectiveness of feedback on energy consumption. A Review for DEFRA of the Literature on Metering, Billing and direct Displays* (Technical report). Environmental Change Institute, University of Oxford. Retrieved from <http://www.eci.ox.ac.uk/research/energy/downloads/smart-metering-report.pdf>
- Diefenbach, N., Cischinsky, H., Rodenfels, M., & Clausnitzer, K.-D. (2010). *Datenbasis Gebäudebestand: Datenerhebung zur energetischen Qualität und zu den Modernisierungstrends im deutschen Wohngebäudebestand*. Darmstadt: Institut Wohnen und Umwelt GmbH. Retrieved from [http://datenbasis.iwu.de/dl/Endbericht\\_Datenbasis.pdf](http://datenbasis.iwu.de/dl/Endbericht_Datenbasis.pdf)
- Edmonds, B. (2010). Bootstrapping knowledge about social phenomena using simulation models. *Journal of Artificial Societies and Social Simulation*, 13(1), 8.
- Epstein, J. M., & Axtell, R. (1996). *Growing Artificial Societies: social science from the bottom up*. MIT Press.
- Ernst, A., & Briegel, R. (2016). A dynamic and spatially explicit psychological model of the diffusion of green electricity across Germany. *Journal of Environmental Psychology*, 52, 183–193.
- Farmer, J. D., & Foley, D. (2009). The economy needs agent-based modelling. *Nature*, 460(7256), 685–686.
- Feuillette, S., Bousquet, F., & Le Goulven, P. (2003). SINUSE: a multi-agent model to negotiate water demand management on a free access water table. *Environmental Modelling & Software*, 18(5), 413–427.
- Friege, J. (2016). Increasing homeowners’ insulation activity in Germany: An empirically grounded agent-based model analysis. *Energy and Buildings*, 128, 756–771.
- Friege, J., & Chappin, E. (2014). Modelling decisions on energy-efficient renovations: a review. *Renewable and Sustainable Energy Reviews*, 39, 196–208.

- Friege, J., Holtz, G., & Chappin, É. J. L. (2016). Exploring Homeowners' Insulation Activity. *Journal of Artificial Societies and Social Simulation*, 19(1), 4. Retrieved from <http://jasss.soc.surrey.ac.uk/19/1/4.html>
- Galán, J.M., Izquierdo, L.R., Izquierdo, S.S., Santos, J.I., Del Olmo, R., López-Paredes, A. & Edmonds, B. (2009). Errors and artefacts in agent-based modelling. *Journal of Artificial Societies and Social Simulation* 12(1), 1. Retrieved from <http://jasss.soc.surrey.ac.uk/12/1/1.html>
- Gilbert, N. (2004). Agent-based social simulation: dealing with complexity. *The Complex Systems Network of Excellence*, 9(25), 1–14.
- Gilbert, N., & Troitzsch, K. G. (2005). *Simulation for the Social Scientist* (2nd ed.). Open University Press.
- Graves, C., Ebbesen, S. D., Mogensen, M., & Lackner, K. S. (2011). Sustainable hydrocarbon fuels by recycling CO<sub>2</sub> and H<sub>2</sub>O with renewable or nuclear energy. *Renewable and Sustainable Energy Reviews*, 15(1), 1–23. <https://doi.org/10.1016/j.rser.2010.07.014>
- Grimm, V. (2005). Pattern-Oriented Modeling of Agent-Based Complex Systems: Lessons from Ecology. *Science*, 310(5750), 987–991. <https://doi.org/10.1126/science.1116681>
- Haarlemmer, G., Boissonnet, G., Peduzzi, E., & Setier, P.-A. (2014). Investment and production costs of synthetic fuels—A literature survey. *Energy*, 66, 667–676.
- Hake, J.-F., Fischer, W., Venghaus, S., & Weckenbrock, C. (2015). The German Energiewende—history and status quo. *Energy*, 92, 532–546.
- Hales, D., Rouchier, J., & Edmonds, B. (2003). Model-to-model analysis. *Journal of Artificial Societies and Social Simulation*, 6(4).
- Hauke, J., Lorscheid, I., & Meyer, M. (2017). Recent Development of Social Simulation as Reflected in JASSS Between 2008 and 2014: A Citation and Co-Citation Analysis. *Journal of Artificial Societies and Social Simulation*, 20(1), 5.
- Heckbert, S., Baynes, T., & Reeson, A. (2010). Agent-based modeling in ecological economics. *Annals of the New York Academy of Sciences*, 1185, 39–53.
- Hedstrom, P. (2005). *Dissecting the Social: On the Principles of Analytical Sociology*. Cambridge University Press.
- Hemmelskamp, J. (1999). *Umweltpolitik und technischer Fortschritt: eine theoretische und empirische Untersuchung der Determinanten von Umweltinnovationen*. Springer-Verlag.
- Holtz, G., Alkemade, F., de Haan, F., Köhler, J., Trutnevyte, E., Luthe, T., Ruutu, S. (2015). Prospects of modelling societal transitions: Position paper of an emerging community. *Environmental Innovation and Societal Transitions*, (17), 41–58. <https://doi.org/10.1016/j.eist.2015.05.006>
- Holzhauser, S., Krebs, F., & Ernst, A. (2013). Considering baseline homophily when generating spatial social networks for agent-based modelling. *Computational and Mathematical Organization Theory*, 19(2), 128–150.
- Janssen, M., & Ostrom, E. (2006). Empirically Based, Agent-based models. *Ecology and Society*, 11(2(37)). Retrieved from <http://www.ecologyandsociety.org/vol11/iss2/art37/>
- Jensen, T., & Chappin, É. J. L. (2017). Reducing domestic heating demand: managing the impact of behavior-changing feedback devices via marketing. *Environmental Management*, 197, 642–655.
- Jensen, T., Holtz, G., Baedeker, C., & Chappin, E. (2016). Energy-efficiency impacts of an air-quality feedback device in residential buildings: An agent-based modeling assessment. *Energy and Buildings*, (116), 151–163.
- Karlin, B., Ford, R., & Squiers, C. (2014). Energy feedback technology: a review and taxonomy of products and platforms. *Energy Efficiency*, 7(3), 377–399.

- Kiesling, E., Günther, M., Stummer, C., & Wakolbinger, L. M. (2012). Agent-based simulation of innovation diffusion: a review. *Central European Journal of Operations Research*, 20(2), 183–230.
- Köhler, J., de Haan, F., Holtz, G., Kubeczko, K., Moallemi, E., Papachristos, G., & Chappin, E. (2018). Modelling Sustainability Transitions: An Assessment of Approaches and Challenges. *Journal of Artificial Societies and Social Simulation*, 21(1), 8. <https://doi.org/10.18564/jasss.3629>
- Krebs, F. (2017). An Empirically Grounded Model of Green Electricity Adoption in Germany: Calibration, Validation and Insights into Patterns of Diffusion. *Journal of Artificial Societies and Social Simulation*, 20(2). [https://doi.org/DOI: 10.18564/jasss.3429](https://doi.org/DOI:10.18564/jasss.3429)
- Laschke, M., Hassenzahl, M., & Diefenbach, S. (2011). Things with attitude: Transformational products (pp. 1–2). Presented at the Create11 conference.
- Latane, B. (1981). The psychology of social impact. *American Psychologist*, 36(4), 343.
- Lauf, T., Gawel, E., & Frank, K. (2015). The spatial allocation of renewable power infrastructure. Presented at the The Computational Social Science Society of the Americas 2015. Retrieved from <https://computationsocialscience.org/csssa2015-conference-schedule/csssa-2015-papers/>
- Liedtke, C., Baedeker, C., Hasselkuß, M., Rohn, H., & Grinewitschus, V. (2015). User-integrated innovation in Sustainable LivingLabs: an experimental infrastructure for researching and developing sustainable product service systems. *Journal of Cleaner Production*, 97, 106–116.
- Macal, C. M., & North, M. J. (2010). Tutorial on agent-based modelling and simulation. *Journal of Simulation*, 4(3), 151–162. <https://doi.org/10.1057/jos.2010.3>
- Mayntz, R. (2004). Mechanisms in the analysis of social macro-phenomena. *Philosophy of the Social Sciences*, 34(2), 237–259.
- Meyer, M., Lorscheid, I., & Troitzsch, K. G. (2009). The Development of Social Simulation as Reflected in the First Ten Years of JASSS: a Citation and Co-Citation Analysis. *Journal of Artificial Societies and Social Simulation*, 12(4(12)). Retrieved from <http://jasss.soc.surrey.ac.uk/12/4/12.html>
- Mielke, J., Vermaßen, H., Ellenbeck, S., Milan, B. F., & Jaeger, C. (2016). Stakeholder involvement in sustainability science—A critical view. *Energy Research & Social Science*, 17, 71–81.
- Moss, S. (2008). Alternative approaches to the empirical validation of agent-based models. *Journal of Artificial Societies and Social Simulation*, 11(1), 5.
- Müller, M., Sensfuß, F., & Wietschel, M. (2007). Simulation of current pricing-tendencies in the German electricity market for private consumption. *Energy Policy*, 35(8), 4283–4294.
- Ormerod, P., & Rosewell, B. (2009). Validation and verification of agent-based models in the social sciences. In *Epistemological Aspects of Computer Simulation in the Social Sciences* (pp. 130–140). Springer.
- Parkhurst, J. (2016). *The politics of evidence: from evidence-based policy to the good governance of evidence*. Routledge.
- Pfenninger, S., Hawkes, A., & Keirstead, J. (2014). Energy systems modeling for twenty-first century energy challenges. *Renewable and Sustainable Energy Reviews*, 33, 74–86.
- Pohl, C., & Hirsch Hadorn, G. (2007). *Principles for designing transdisciplinary research. Proposed by the Swiss Academies of Arts and Sciences*. Munich: Oekom Verlag.
- ProClim-. (1997). *Research on sustainability and global change - visions in science policy by swiss researchers* (Visions by Swiss Researchers). ProClim - Forum for Climate and Global Change and CASS - Conference of the Swiss Scientific Academies. Retrieved from <https://naturalsciences.ch/service/publications/75640-visions-of-swiss-scientists>



- Quitow, R., Roehrkasten, S., & Jaenicke, M. (2016). *The German Energy Transition in International Perspective* (IASS Study). Potsdam, Germany: Institute for Advanced Sustainability Studies (IASS) e. V. <https://doi.org/10.2312/iass.2016.009>
- Rai, V., & Henry, A. D. (2016). Agent-based modelling of consumer energy choices. *Nature Climate Change*, 6(6), 556–562. <https://doi.org/10.1038/nclimate2967>
- Reeg, M., Hauser, W., Wassermann, S., Kast, T., Klann, U., Nienhaus, K., Weimer-Jehle, W. (2012). Amiris: An agent-based simulation model for the analysis of different support schemes and their effects on actors involved in the integration of renewable energies into energy markets (pp. 339–344). Presented at the Database and Expert Systems Applications (DEXA), 2012 23rd International Workshop on, IEEE.
- Ringler, P., Keles, D., & Fichtner, W. (2016). Agent-based modelling and simulation of smart electricity grids and markets—a literature review. *Renewable and Sustainable Energy Reviews*, 57, 205–215.
- Scheidt, M. (2002). *Ein Modell zur Mikrosimulation des Spothandels von Strom auf der Basis eines Multi-Agenten-Systems*. Rheinisch-Westfälische Technische Hochschule Aachen, Aachen.
- Schenuit, C., Heuke, R., & Paschke, J. (2016). *Potentialatlas Power to Gas. Klimaschutz umsetzen, erneuerbare Energien integrieren, regionale Wertschöpfung ermöglichen*. Berlin: Deutsche Energie - Agentur GmbH.
- Schneidewind, U., & Scheck, H. (2013). Die Stadt als „Reallabor“ für Systeminnovationen. In *Soziale Innovation und Nachhaltigkeit* (pp. 229–248). Springer.
- Schnülle, C., Kisjes, K., Thier, P., Stührmann, T., Nikolic, I., von Gleich, A., & Gößling-Reisemann, S. (2016). Optionen für die Integration von Power-to-Fuel in den Energiewendeprozess aus einer sozioökonomischen Perspektive. *Vierteljahreshefte Zur Wirtschaftsforschung. DIW Berlin*, (85).
- Schot, J., Kanger, L., & Verbong, G. (2016). The roles of users in shaping transitions to new energy systems. *Nature Energy*, 1(5), 16054. <https://doi.org/10.1038/nenergy.2016.54>
- Schubert, D. K. J., Thuß, S., & Möst, D. (2015). Does political and social feasibility matter in energy scenarios? *Energy Research & Social Science*, 7, 43–54.
- Schwarz, N., & Ernst, A. (2009). Agent-based modeling of the diffusion of environmental innovations — An empirical approach. *Technological Forecasting and Social Change*, 76(4), 497–511. <https://doi.org/10.1016/j.techfore.2008.03.024>
- Sensfuß, F., Ragwitz, M., & Genoese, M. (2008). The merit-order effect: A detailed analysis of the price effect of renewable electricity generation on spot market prices in Germany. *Energy Policy*, 36(8), 3086–3094.
- Sensfuß, F., Ragwitz, M., Genoese, M., & Möst, D. (2007). *Agent-based simulation of electricity markets: a literature review*. Working paper sustainability and innovation. Retrieved from <http://hdl.handle.net/10419/28520>
- Sorda, G., Sunak, Y., & Madlener, R. (2013). An agent-based spatial simulation to evaluate the promotion of electricity from agricultural biogas plants in Germany. *Ecological Economics*, 89, 43–60.
- Sovacool, B. K. (2014). What are we doing here? Analyzing fifteen years of energy scholarship and proposing a social science research agenda. *Energy Research & Social Science*, 1, 1–29.
- Steg, L., Perlaviciute, G., & van der Werff, E. (2015). Understanding the human dimensions of a sustainable energy transition. *Frontiers in Psychology*, 6. <https://doi.org/10.3389/fpsyg.2015.00805>
- Tremel, A., Wasserscheid, P., Baldauf, M., & Hammer, T. (2015). Techno-economic analysis for the synthesis of liquid and gaseous fuels based on hydrogen production via electrolysis. *International Journal of Hydrogen Energy*, 40(35), 11457–11464.

- Voinov, A., & Bousquet, F. (2010). Modelling with stakeholders. *Environmental Modelling & Software*, 25(11), 1268–1281. <https://doi.org/10.1016/j.envsoft.2010.03.007>
- Wagner, F., & Grunwald, A. (2015). Reallabore als Forschungs-und Transformationsinstrument Die Quadratur des hermeneutischen Zirkels. *GAIA-Ecological Perspectives for Science and Society*, 24(1), 26–31.
- Weidlich, A., Sensfuß, F., Genoese, M., & Veit, D. (2008). Studying the effects of CO2 emissions trading on the electricity market: A multi-agent-based approach. In *Emissions Trading* (pp. 91–101). Springer.
- Windrum, P., Fagiolo, G., & Moneta, A. (2007). Empirical validation of agent-based models: Alternatives and prospects. *Journal of Artificial Societies and Social Simulation*, 10(2), 8.
- Wood, G., & Newborough, M. (2003). Dynamic energy-consumption indicators for domestic appliances: environment, behaviour and design. *Energy and Buildings*, 35(8), 821–841.
- Yadack, M., Bogner, K., Vermeulen, B., Graebig, M., & Pyka, A. (2016). Stadtwerke als Gestalter der Energiewende. Regionalität ist klarer Wettbewerbsvorteil. *Emw – Energie.Markt.Wettbewerb.*, October.
- Yadack, M., Vermeulen, B., & Pyka, A. (2016a). Competition in the German market for retail electricity: An agent-based simulation. In B. Vermeulen & M. Paier (Eds.), *Innovation Networks for Regional Development. Concepts, Case studies, and Agent-based Models*. Springer.